



# Syed Mamun Raihan

LEAD ENGINEER

# Summary of Skills

- ▶ 15 years of industry and research experiences, C/C++ is core expertise
- ▶ *Object oriented design, meta-programming, design patterns, Boost/STL*
- ▶ Motivated, passionate, adaptable
- ▶ Uses Scrum/Agile/XP methodologies
- ▶ Comfortable with gdb/gcc and Visual Studio
- ▶ Brain bench C++ Ranking 88%, score 3.88 (#8120314)
- ▶ Working knowledge of SQL 92, Object Oriented Databases
- ▶ Working knowledge of Python, Matlab, UML 2.0, Tcl/Tk

# Summary of Work and Education

- ▶ B.Sc. in EEE from Bangladesh University of Engineering & Technology
- ▶ M.Sc. in Computer Engineering from University of Central Florida
- ▶ 2 years of PhD Research and Study (incomplete) from University of Western Ontario
- ▶ 4/2009 to Current – Working for Ciena Inc. (Ciena is a major optical networking supplier for ATT and Verizon). Working in Software Defined Networking. My presentation involves my work related with Open flow.

# Experiences (Highlights)

- ▶ 4/2008 to 2/2009 – Worked for Algorithmics Inc. (a IBM company). Worked in Basel 1 and Basel 2 portfolio risk assessment tools based on Risk++ (a C++ component library developed for financial industry including central banks).
- ▶ 5/2007 to 2/2008 – Worked for Nortel (Enterprise Division). Designed and developed pro-active voice quality monitoring systems. Used visual studio for windows client development and phone client using JTAG.
- ▶ 1/2003 to 4/2004 – Worked for North South University as a Lecturer. Taught C++/Programming Languages, Advance Database concepts in undergraduate classes.
- ▶ 5/2000 to 11/2001 – Worked for Spike Broadband Inc.. Worked on SNMP/CLI management application using C.

# Software defined networking

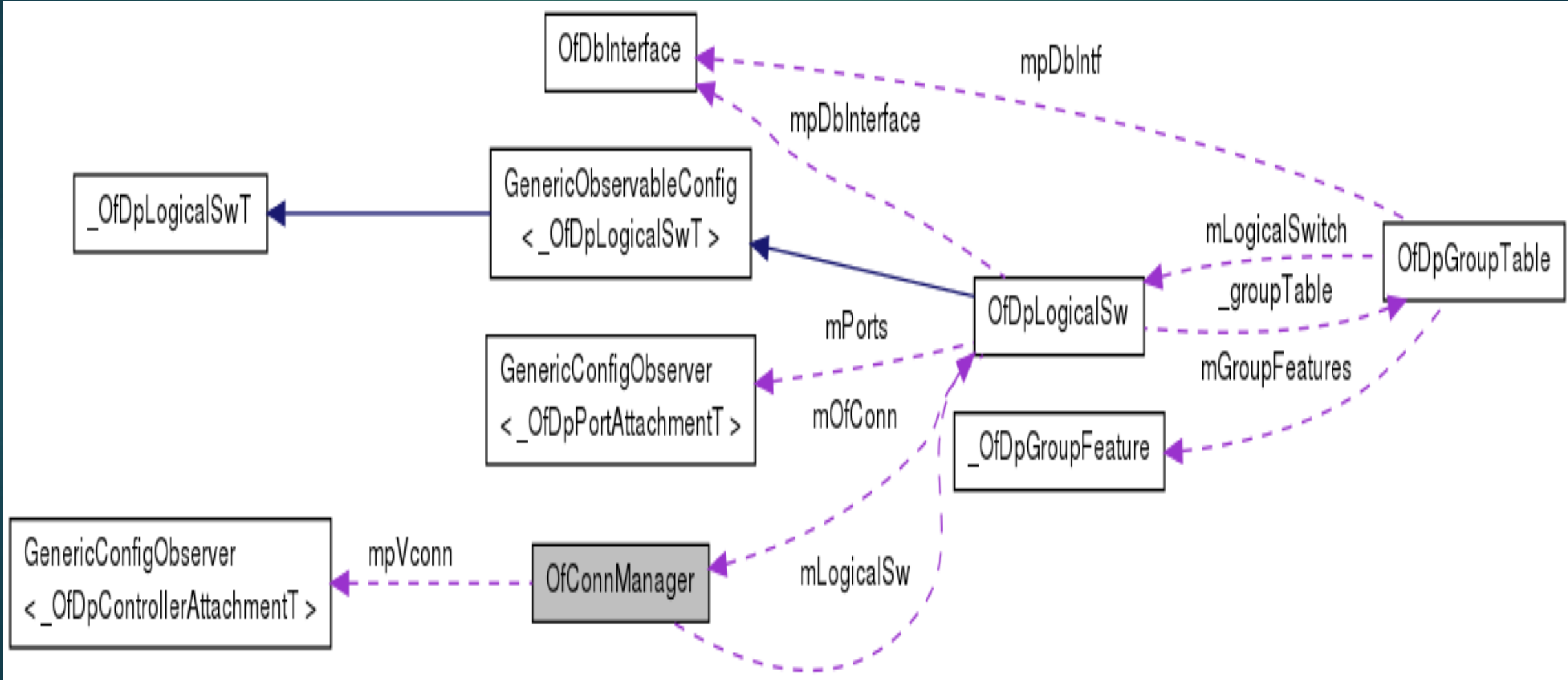
- ▶ SDN approach in Ciena is Open flow driven. See (<https://www.opennetworking.org/>)
- ▶ Open Networking Foundation (ONF) is a user-driven organization dedicated to the promotion and adoption of [Software-Defined Networking \(SDN\)](#) through open standards development.
- ▶ Open flow Standard Specification covers the components and the basic functions of the network switch, and the OpenFlow protocol to manage an OpenFlow switch from a remote controller.
- ▶ I have implemented network front end for a switch and controller.
- ▶ I have contributed specification and platform independent switch core logic to support Flows, Groups, Queues.
- ▶ I have implemented Open flow switch configuration API and contributed Rest API interface to OF-CONFIG-API.

# OF Switch Design Considerations

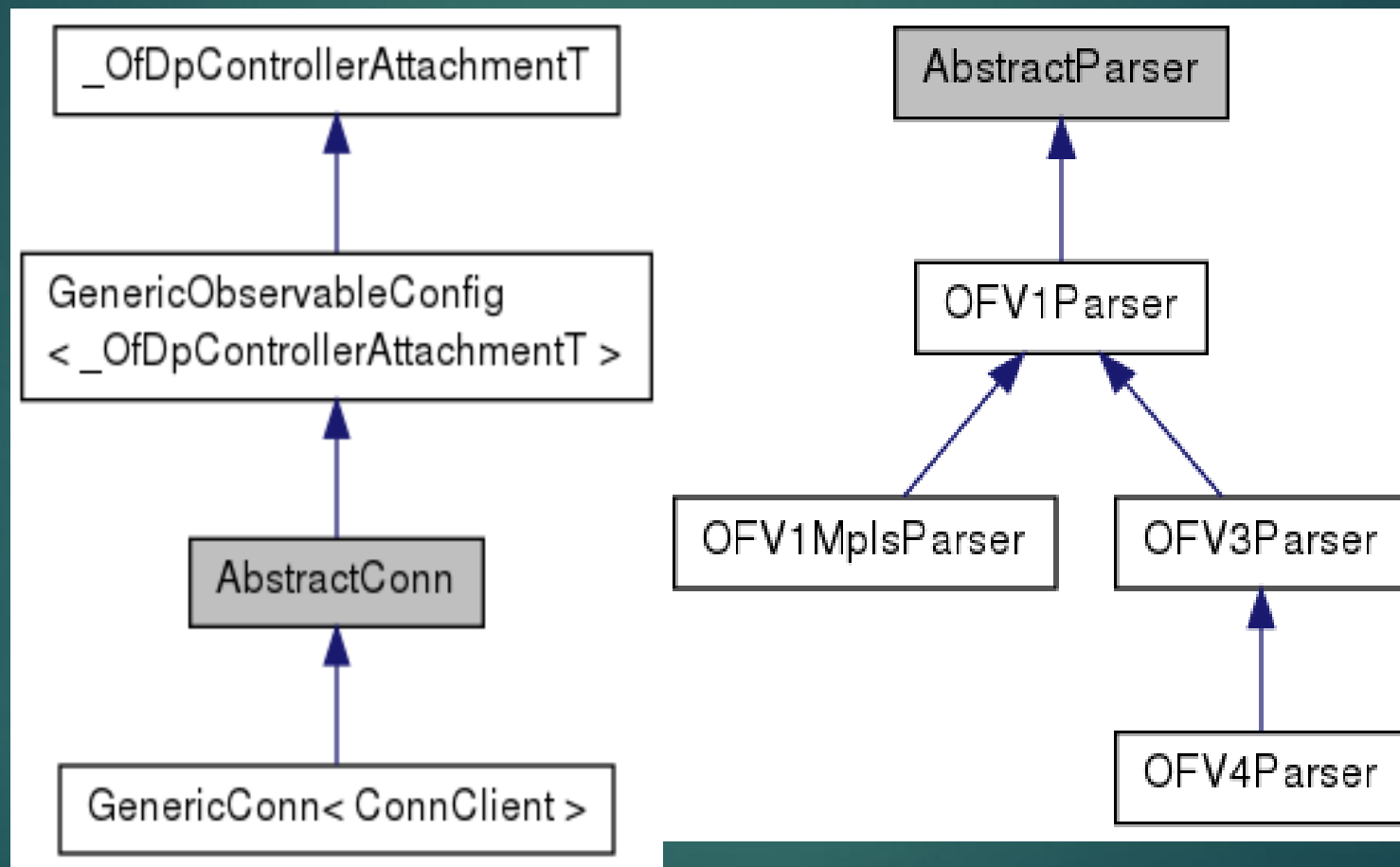
- ▶ Must support multiple OF Spec. Currently officially supports 1.0 and 1.3.
- ▶ Must support multiple connection mechanism. Currently support TLSv1 and TCP connections.
- ▶ Must support alternate configuration mechanism. Work in progress.
- ▶ Data plane API must support multiple HW platforms. Work in progress.
- ▶ Observer, Decorator, Proxy, Singleton, State, Virtual Constructor, Factory Method, Memento and Visitor patterns have been used or work in progress
- ▶ Doxygen Documents available for code walk.



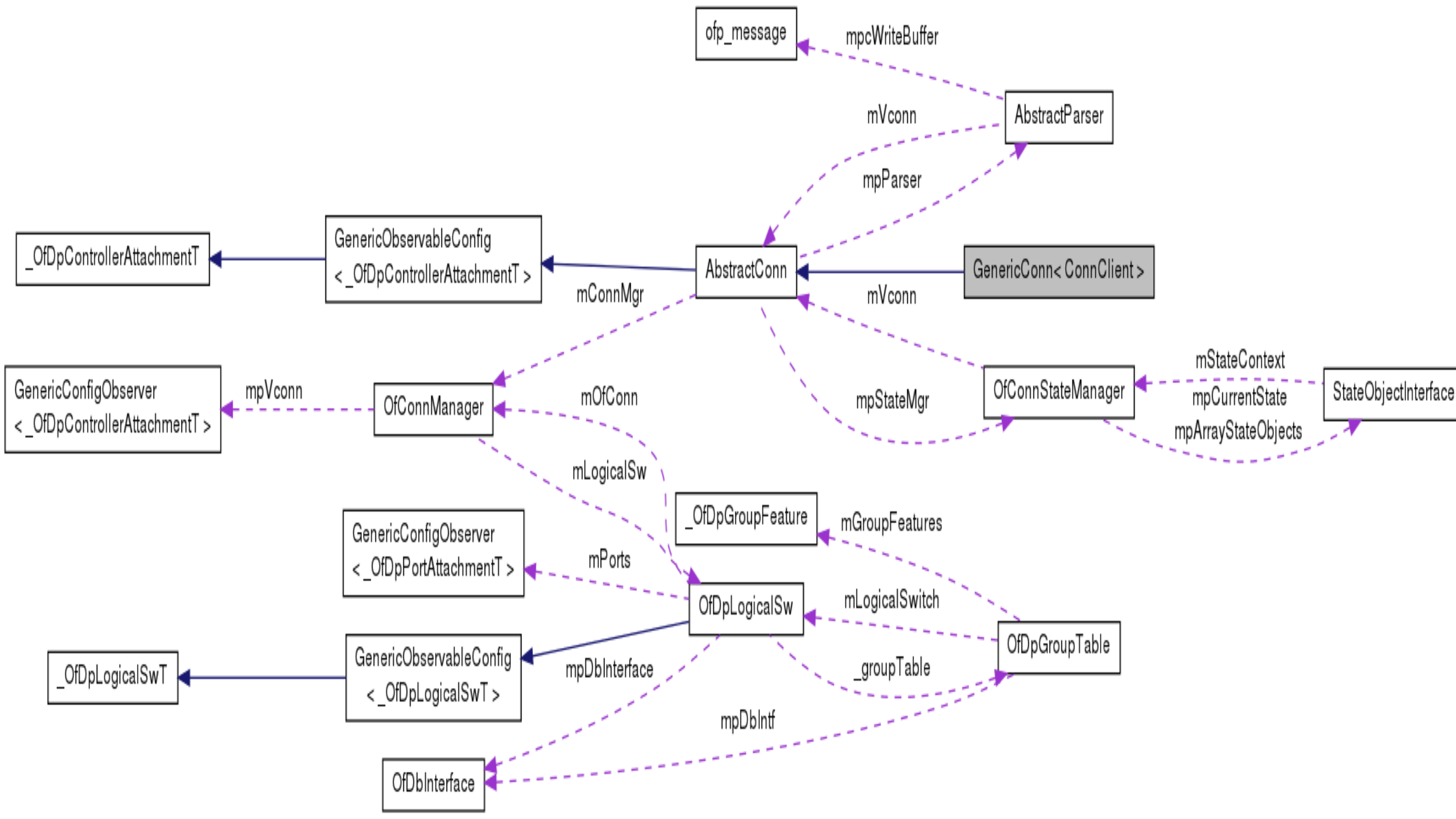
# Open flow Connection Manager



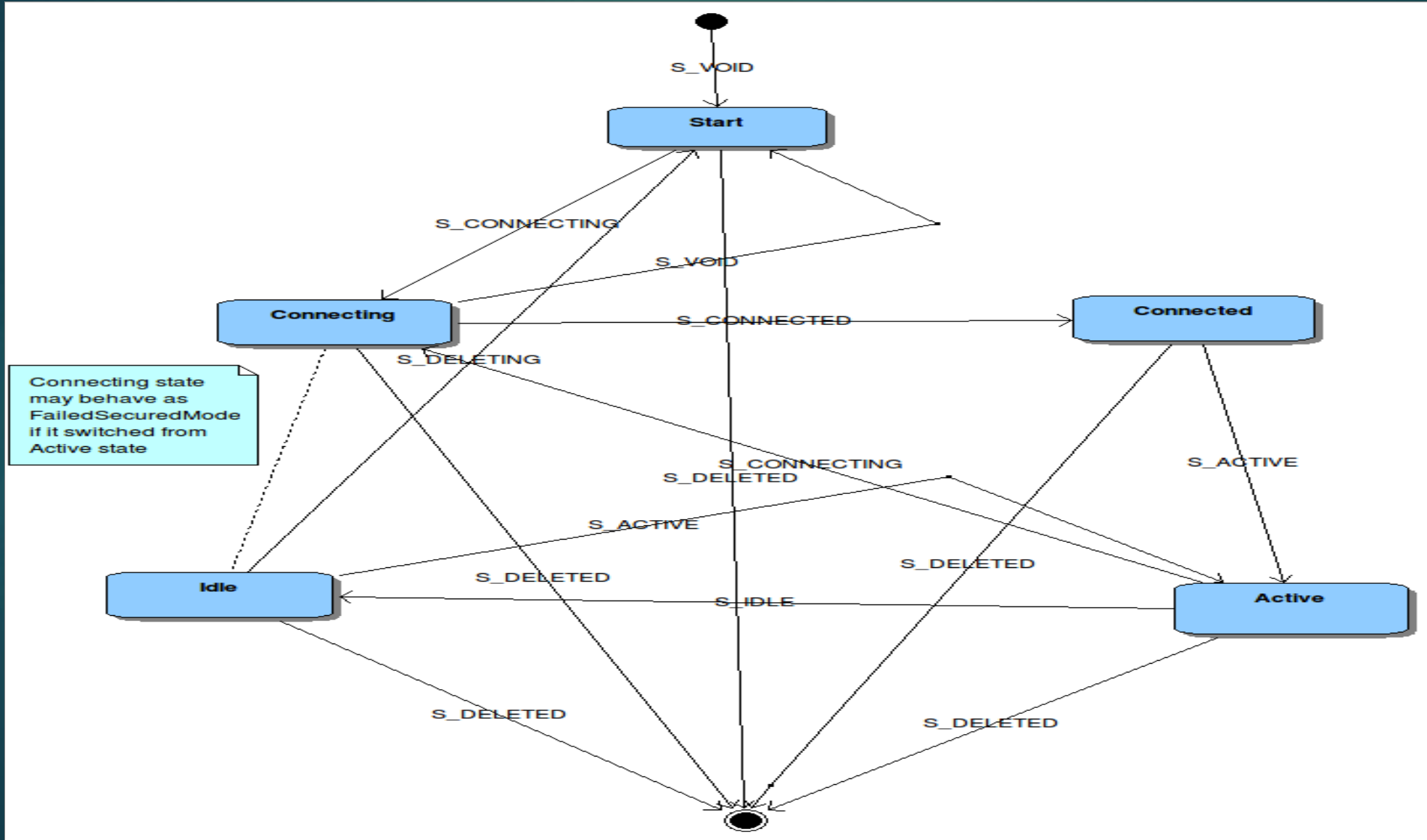
# Abstract Connection / Parser



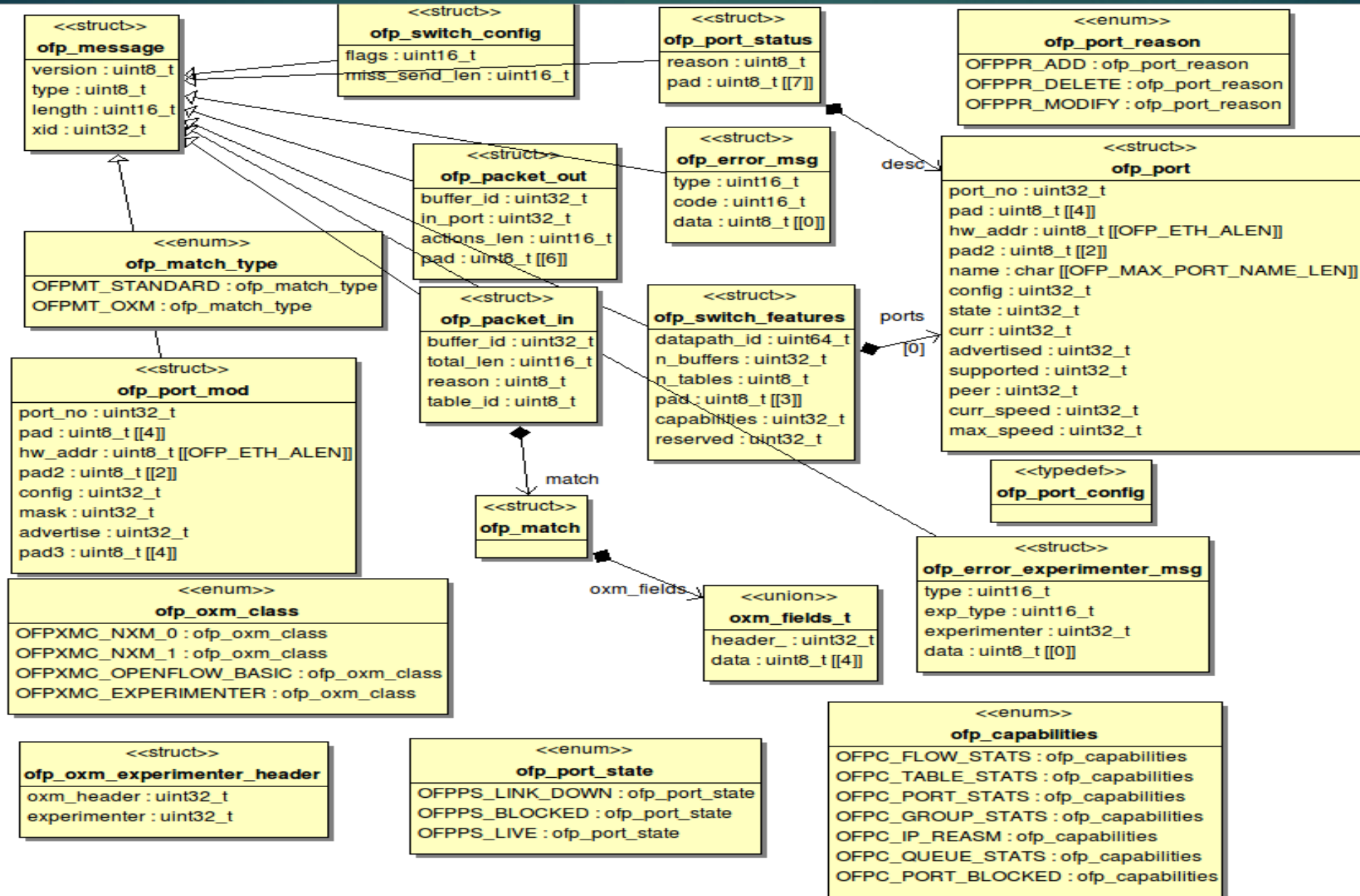




# Connection State Machine

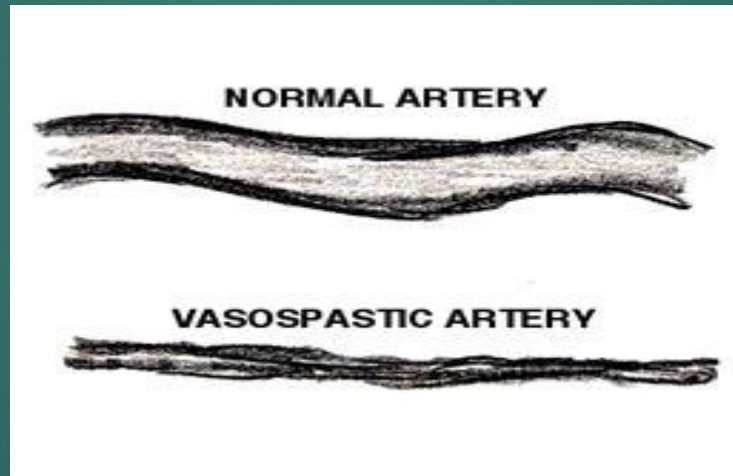


# Sample external data model for switch and controller



# Detection of Vasospasm in Comatose Patients Using EEG Spectral Characteristics (C++/matlab)

- ▶ About 30,000 cases of Subarachnoid Hemorrhage<sup>‡</sup>
  - ▶ 20,000 develop detectable vasospasm
  - ▶ 10,000 will become symptomatic
  - ▶ 60% suffer death or disability



- ▶ Treatment is possible if detected early

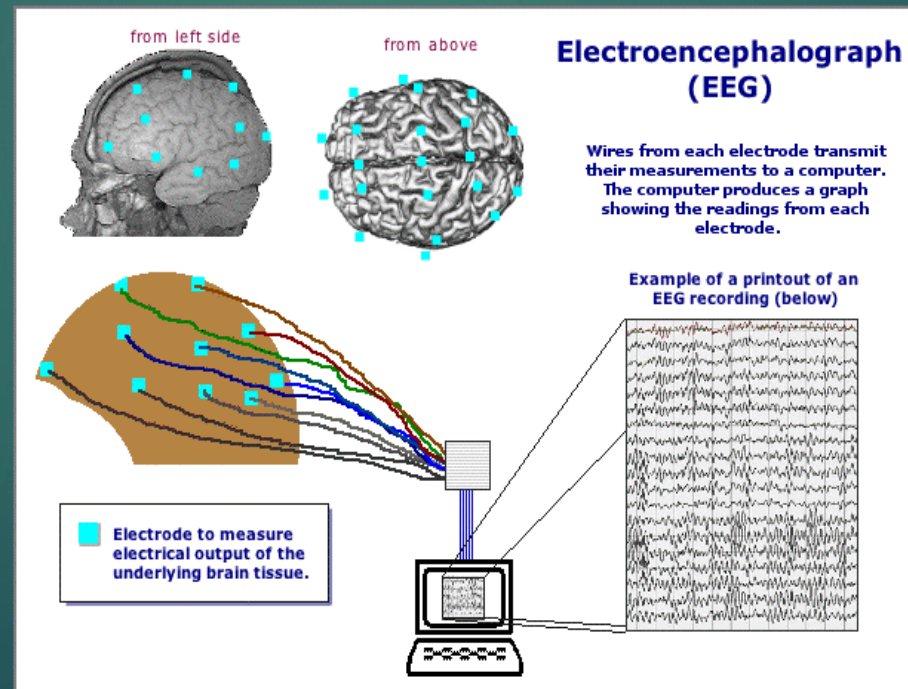
<sup>‡</sup> Heiserman JE, "MR Angiography for the Diagnosis of Vasospasm after Subarachnoid Hemorrhage. Is It Accurate? Is It Safe?", American Journal of Neuroradiology 21:1571-1572 (2000)

## - CURRENT MODES OF DETECTION -

- ▶ Transcranial Doppler or angiography are used
  - ▶ Neurological symptoms appear late
  - ▶ May result in late detection
- ▶ Current Role of EEG in monitoring vasospasm
  - ▶ Can record EEG continuously for days
  - ▶ Qualitative analysis of EEG requires specialized skills
  - ▶ May result in late detection

## - OBJECTIVE -

- ▶ To design and evaluate an algorithm for automatic detection of vasospasm from EEG data
  - ▶ using quantitative features extracted from EEG spectral characteristics
  - ▶ using autoregressive (AR) modeling





## - AUTOREGRESSIVE (AR) MODELLING -

- ▶ AR model is based on parametric modelling in which each value in a time series is predicted from the past values

$$x_t = -\sum_{m=1}^p A_m x_{t-m} + e_t$$

where  $x_t$  is time series at time  $t$ ,

$x_{t-m}$  is the  $m^{th}$  previous value,

$A_m$  are the AR model coefficients, and

$e_t$  is zero mean white noise



## - HIGH RESOLUTION SPECTRUM -

- Spectral power estimation using AR model

$$S_f = \frac{\sigma^2 T}{\left| \sum_{m=0}^p A_m e^{(-j2\pi fm)} \right|^2}$$

where  $S_f$  is power at frequency  $f$ ,

$\sigma^2$  is the variance for a particular model order  $p$ ,

$T$  is the length of the time series, and

$A_m$  is the  $m^{th}$  coefficients of the time series

## – SPECTRAL FLATNESS RATIO –

- ▶ SFR is calculated using the following equation in the Alpha band

$$SFR = \frac{e^{((\sum \log(S_f))/N)}}{(\sum S_f)/N}$$

- ▶ Numerator is the geometric mean of the power spectrum (expressed in dB), and
- ▶ Denominator is the arithmetic mean of the power spectrum (expressed in dB)

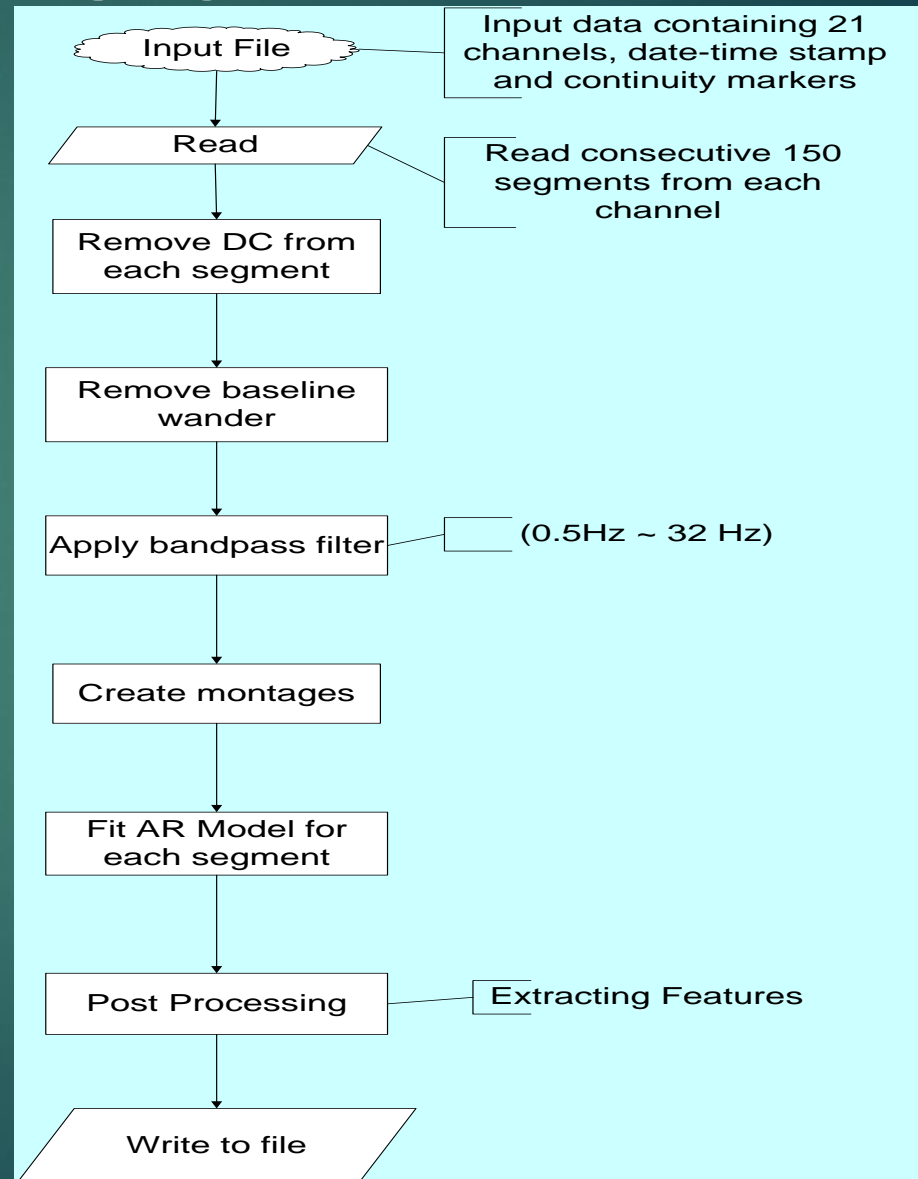
# - FEATURE EXTRACTION -

## ► Pre-processing

- Data are read and segmented
- DC is removed
- Trend is removed
- Data are filtered
- Montage created

## ► Post-processing

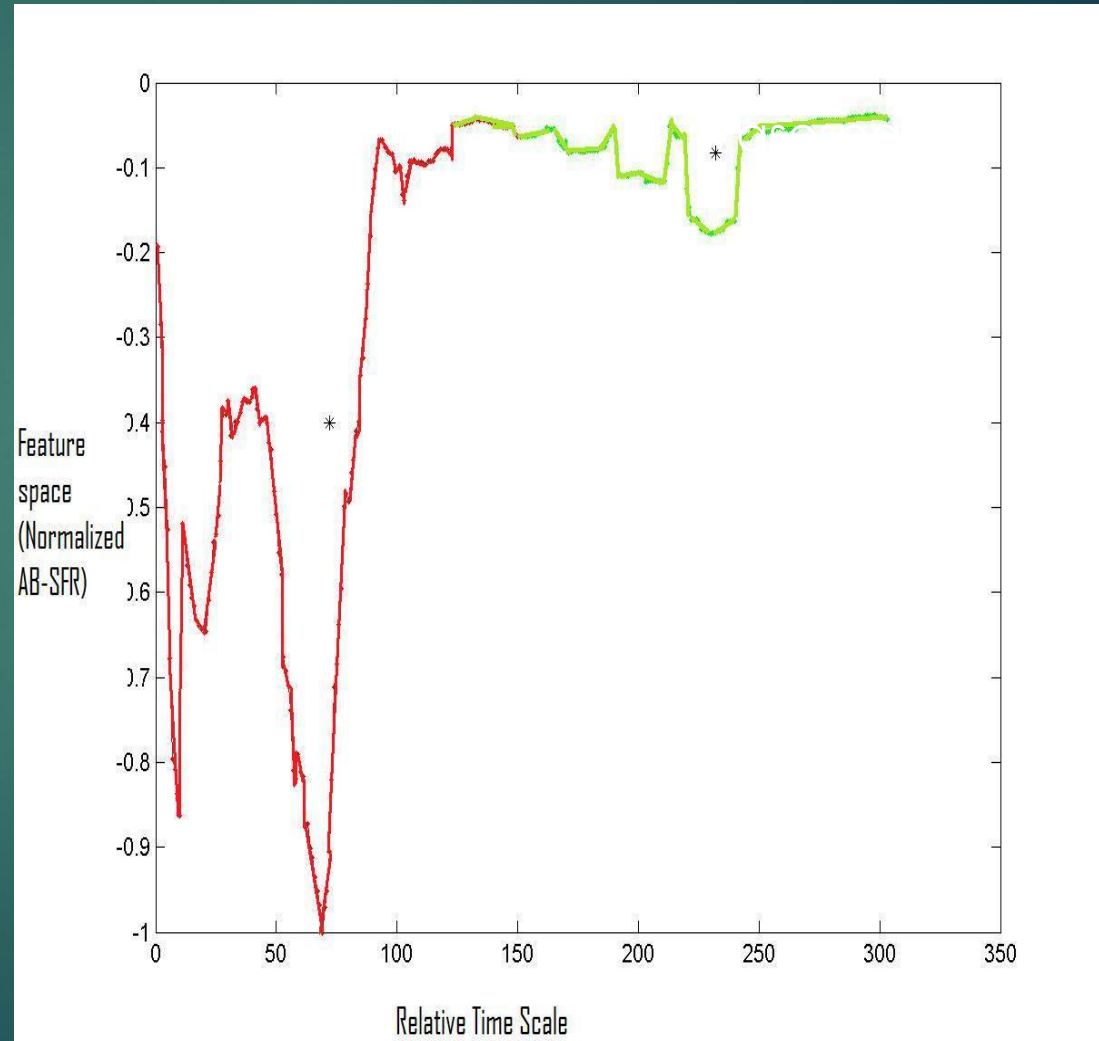
- AR model is fitted
- AB-SFR is calculated
- Result is saved



## - CLASSIFICATION -

- ▶ Fuzzy C-mean clustering is used
- ▶ Maximum cluster distance in feature space is used to identify montage
- ▶ Class closer to 0 dB is used to detect onset of vasospasm

Clustered data of a sample montage



## - EVALUATION -

- ▶ Goodness of fit using Chi-Squared Test
  - ▶ Uses *TPF* and *TNF*
- ▶ Test for existence of two groups
  - ▶ Uses *Students-T test* on *AB-SFR*
- ▶ Margin of Error

$$E = Z\alpha \sqrt{\frac{p(1-p)}{n}} \quad \dagger$$

where

$n$  is the sample size,

$p$  is the classification accuracy, and

$Z\alpha = 1.96$  at 95% confidence interval

<sup>†</sup> Triola MM, Triola MF, Biostatistics for the biological and health sciences, Pearson Addison Wesley, Boston, 261, (2006)

## - RESULTS -

- ▶ Classification results in identifying 10/12 vasospasm segments and 10/12 non-vasospasm segments
  - ▶  $\chi^2_{calculated}$  with Yates correction is 0.45 is less than  $\chi^2_{critical}$  (3.84). Therefore classification accuracy is good
  - ▶ Students-T test on AB-SFR is 7.089 when  $t_{critical} = 2.008$  (2-tailed at  $p=0.05$ ). Therefore AB-SFRs are from two different groups
  - ▶ Margin of error is 30%

## - DISCUSSION -

- ▶ Margin of error will be reduced as more patient data become available
- ▶ Cepstrum, AR coefficients and composite features may improve quality of classification
- ▶ Improve algorithm 'intelligence'
  - ▶ Using clinical data like age, sex, neurological symptoms
  - ▶ Continuous monitoring may replace need for TCD



## - CONCLUSION -

- ▶ Quantitative analysis to detect onset of vasospasm
  - ▶ Algorithm may be integrated with any EEG system
  - ▶ Automatic warning system may be implemented
- ▶ Simple non-adaptive high resolution spectral analysis may be sufficient

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